

Complexity Theory

Lecture 10

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<http://www.cl.cam.ac.uk/teaching/2324/Complexity>

One Way Functions

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1. f is one-to-one.

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4. f^{-1} is *not* computable in polynomial time.

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One-way functions exist *if, and only if*, $P \neq UP$.

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Define the function f_U by

*if x is a string that encodes an accepting computation of U ,
then $f_U(x) = 1y$ where y is the input string accepted by this
computation.*

$f_U(x) = 0x$ otherwise.

We can prove that f_U is a one-way function.

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$\text{NSPACE}(f)$ is the class of languages accepted by a *nondeterministic* Turing machine using at most $O(f(n))$ work space.

As we are only counting work space, it makes sense to consider bounding functions f that are less than linear.

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Moreover,

$$L \subseteq NL \cap co-NL$$

$$P \subseteq NP \cap co-NP$$

$$PSPACE \subseteq NPSPACE \cap co-NPSPACE$$

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Thus $S \in PSPACE$.

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Definition

A function $f : \mathbb{N} \rightarrow \mathbb{N}$ is *constructible* if:

- f is non-decreasing, i.e. $f(n+1) \geq f(n)$ for all n ; and
- there is a deterministic machine M which, on any input of length n , replaces the input with the string $0^{f(n)}$, and M runs in time $O(n + f(n))$ and uses $O(f(n))$ *work space*.

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If f and g are constructible functions, then so are $f + g$, $f \cdot g$, 2^f and $f(g)$ (this last, provided that $f(n) > n$).

Using Constructible Functions

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If f is a constructible function then any language in $\text{NTIME}(f)$ is accepted by a machine for which all computations are of length at most $O(f(n))$.

Also, given a Turing machine M and a constructible function f , we can define a machine that simulates M for $f(n)$ steps.

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The first two are straightforward from definitions.

The third is an easy simulation.

The last requires some more work.

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2. while S is not empty, choose node i in S : remove i from S and for all j such that there is an edge (i, j) and j is unmarked, mark j and add j to S ;
3. if b is marked, accept else reject.

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Let M be a nondeterministic machine working in space bounds $f(n)$.

For any input x of length n , there is a constant c (depending on the number of states and alphabet of M) such that the total number of possible configurations of M within space bounds $f(n)$ is bounded by $n \cdot c^{f(n)}$.

Here, $c^{f(n)}$ represents the number of different possible contents of the work space, and n different head positions on the input.

Questions?